## 3D <br> acquisition

Computer Vision

## 3D acquisition taxonomy



Computer Vision

## 3D acquisition taxonomy



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Stereo

The underlying principle is "triangulation" :


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## (Passive) stereo

## Simple configuration :



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q identical cameras
q coplanar image planes
$q$ aligned $x$-axes

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## A simple stereo setup

## Reminder :


the camera projection can be formulated as

$$
\rho p=K R^{t}(P-C)
$$

for some non-zero $\rho \in \mathbb{R}$
Here $R$ is the identity...

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A simple stereo setup


$$
\left.\rho\left(\begin{array}{l}
x \\
y \\
1
\end{array}\right)=K\left(\begin{array}{l}
X \\
Y \\
Z
\end{array}\right)\left|\left(\begin{array}{l}
x^{\prime} \\
\rho^{\prime} \\
y^{\prime} \\
1
\end{array}\right)=K\left(\begin{array}{l}
X-b \\
Y \\
Z
\end{array}\right)\right| \begin{array}{ccc}
f k_{x} 0 & 0 \\
0 & f k_{y} & 0 \\
0 & 0 & 1
\end{array}\right)
$$

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A simple stereo setup


$$
\left\{\begin{array} { l } 
{ x = \frac { f k _ { x } X } { Z } , } \\
{ y = \frac { f k _ { y } Y } { Z } , }
\end{array} \text { and } \left\{\begin{array}{l}
x^{\prime}=\frac{f k_{x}(X-b)}{Z}, \\
y^{\prime}=\frac{f k_{y} Y}{Z},
\end{array}\right.\right.
$$

Note that $y=y^{\prime}$

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## A simple stereo setup

The 3D coordinates of the point are

$$
\begin{aligned}
X & =b \frac{x}{\left(x-x^{\prime}\right)} \\
Y & =b \frac{k_{x}}{k_{y}} \frac{y}{\left(x-x^{\prime}\right)} \\
Z & =b k_{x} \frac{f}{\left(x-x^{\prime}\right)}
\end{aligned}
$$

$\left(x-x^{\prime}\right)$ is the so-called disparity
Stereo is imprecise for far away objects, but increasing $b$ and/or $f$ can increase depth resolution

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## A simple stereo setup

Notice: for this simple setup, same disparity means same depth


Computer Vision
same disparity means same depth


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## A simple stereo setup

Increasing $b$ increases depth resolution


one has to strike a balance with visibility...

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## A simple stereo setup

Increasing $f$ increases depth resolution

one has to strike a balance with visibility...

## Remarks

$r$ 1. increasing $b$ and/or $f$ increases depth resolution but reduces simultaneous visibility
r 2. iso-disparity loci are depth planes, not so for other configurations
r 3. as soon as the disparity gets too small, depth difference can no longer be seen; hence human stereo only works up to $\pm 10 \mathrm{~m}$
$r$ 4. the real problem is finding correspondences

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A simple stereo setup


The HARD problem is finding the correspondences

Notice : no reconstruction for the untextured back wall...

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A simple stereo setup


The HARD problem is finding the correspondences

Notice : no reconstruction for the untextured back wall...

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Computer Vision


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## Stereo, the general setup

We start by the relation between the two projections of a point, to ease correspondence search
In the second image the point must be along the projection of the viewing ray for the first camera :


## Stereo, the general setup

We cast this constraint in mathematical expressions :
$p$ and $p^{\prime}$ are the two images of $P$

$$
\begin{aligned}
& \mu p=K R^{t}(P-C) \\
& \rho^{\prime} p^{\prime}=K^{\prime} R^{\prime t}\left(P-C^{\prime}\right)
\end{aligned}
$$

w.r.t. world frame $P$ is on the ray with equation

$$
P=C+\mu R K^{-1} p \quad \text { for some } \mu \in \mathbb{R}
$$

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## Stereo, the general setup

so, the ray is given by

$$
P=C+\mu R K^{-1} p \text { for some } \mu \in \mathbb{R}
$$

Now we project it onto the second image In general, points project there as follows :

$$
\rho^{\prime} p^{\prime}=K^{\prime} R^{\prime t}\left(P-C^{\prime}\right)
$$

and thus, filling in the ray's equation
$\rho^{\prime} p^{\prime}=\mu K^{\prime} R^{\prime t} R K^{-1} p+K^{\prime} R^{\prime t}\left(C-C^{\prime}\right)$

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 VisionStereo, the general setup
the projected ray was found to be
$\rho^{\prime} p^{\prime}=\mu K^{\prime} R^{\prime t} R K^{-1} p+K^{\prime} R^{\prime t}\left(C-C^{\prime}\right)$
the second term is the projection of the 1 st camera's center, the so-called epipole

$$
\rho_{e}^{\prime} e^{\prime}=K^{\prime} R^{\prime t}\left(C-C^{\prime}\right)
$$

the first term is the projection of the ray's point at infinity, the so-called vanishing point finally, adopting the simplifying notation
$A=\quad K^{\prime} R^{\prime t} R K^{-1}$

$$
\rho^{\prime} p^{\prime}=\mu A p+\rho_{e}^{\prime} e^{\prime}
$$

$A$ is the infinity homography

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## Stereo, the general setup

the projected ray

$$
\begin{gathered}
\rho^{\prime} p^{\prime}=\mu K^{\prime} R^{\prime t} R K^{-1} p+K^{\prime} R^{\prime t}\left(C-C^{\prime}\right) \\
\text { or } \\
\rho^{\prime} p^{\prime}=\rho_{e}^{\prime}\left(\mu A p+e^{\prime}\right) \\
\text { is called the epipolar line for } p
\end{gathered}
$$

and runs through the points $A p$ and $e^{\prime}$

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## Stereo, the general setup

note that the epipole lies on all the epipolar lines


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## Stereo, the general setup

$$
\rho^{\prime} p^{\prime}=\mu A p+\rho_{e}^{\prime} e^{\prime}
$$

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## Stereo, the general setup

$$
\rho^{\prime} p^{\prime}=\mu A p+\rho_{e}^{\prime} e^{\prime}
$$

expresses that $p$ 'lies on the line $l$ 'through the epipole $e^{\prime}$ and the vanishing point $A p$ of the ray of sight of $p$ (in the $2^{n d}$ image)

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## Stereo, the general setup

$$
\rho^{\prime} p^{\prime}=\mu A p+\rho_{e}^{\prime} e^{\prime}
$$

the epipolar constraint (epipolar line)
we can rewrite this constraint as

$$
\left|p^{\prime} e^{\prime} A p\right|=p^{\prime \prime}\left(e^{\prime} \times A p\right)=0
$$

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Stereo, the general setup

$$
\left|p^{\prime} e^{\prime} A p\right|=p^{\prime \prime}\left(e^{\prime} \times A p\right)=0
$$

can be written, given

$$
\begin{aligned}
& \quad\left[e^{\prime}\right]_{\times}=\left(\begin{array}{rrr}
0 & -e_{3}^{\prime} & e_{2}^{\prime} \\
e_{3}^{\prime} & 0 & -e_{1}^{\prime} \\
-e_{2}^{\prime} & e_{1}^{\prime} & 0
\end{array}\right) \\
& \text { as } \\
& \left|p^{\prime} e^{\prime} A p\right|=
\end{aligned} p^{\prime t}\left[e^{\prime}\right]_{\times} A p-1 .
$$

$F=\left[e^{\prime}\right]_{\times} A$ is the fundamental matrix
$F$ is a $3 \times 3$ matrix, but has rank 2

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## Stereo, the general setup

$$
p^{\prime t}[e]_{\times} A p=0 \rightarrow p^{p^{\prime} F p=0}
$$

The 3-vector $p^{\prime}{ }^{t} F$ contains the line coordinates of the epipolar line of $p^{\prime}$ (i.e. a line in the 1 st image that contains its corresponding point $p$ )

The 3-vector $F p$ contains the line coordinates of the epipolar line of $p$ (i.e. a line in the 2nd image that contains its corresponding point $p^{\prime}$ )

Hence, the epipolar matrix works in both directions

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Stereo, the general setup


Andrea Fusiello, CVonline

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Epipolar geometry cont'd


## Computer Vision

## Epipolar geometry cont'd

- Epipolar lines are in mutual correspondence

- allows to separate matching problem: matching pts on an epipolar line to pts on the corresponding epipolar line

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## Exploiting epipolar geometry

## Separate 2D correspondence search problem to 1D

 search problem by using two view geometry

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Epipolar geometry cont'd


## Computer

 Vision
## Stereo, the general setup

q one point yields one equation $p^{\prime t} F p=0$ that is linear in the entries of the fundamental matrix $F$
so, we can actually obtain $F$ without any prior knowledge about camera settings if we have sufficient pairs of corresponding points !!
q F can be computed linearly from 8 pairs of corresponding points, i.e. already from 8 'correspondences’ (not 9 , as this is a homogeneous system and one coefficient can be fixed to value 1 to fix the scale !)
q F being rank 2 yields an additional, but non-linear constraint. Thus, 7 correspondences suffice to non-linearly solve for $F$

Stereo, the general setup

## Remarks :

q Of course, in practice one wants to use as many correspondences as available, e.g. for obtaining a least-squares solution, based on the linear system, followed by a step to impose rank 2.
q Often, F is found through `RANSAC' (RANdom Sample Consensus), a procedure to fend off against correspondences that are wrong ('outliers'). It starts from a randomly drawn subset of correspondences of minimal size (e.g. 8), and then keeps on drawing until a subset is found that yields an F so that many correspondences are seen to obey the epipolar constraint. Consistent correspondences (inliers) are then used to refine the solution for $F{ }^{41}$

## Relations between 3 views

one could use more than 2 images, e.g. 3 suppose $P$ projects to $p, p$, and $p$ "
$p$ " is found at the intersection of epipolar lines :

fails when the epipolar lines coincide

$$
\Rightarrow \quad \text { trifocal constraints }
$$

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## Relations between 3 views



## Correspondence problem : constraints

Reducing the search space :
$n$ 1. Points on the epipolar line
$n$ 2. Min. and max. depth $\Rightarrow$ line segment
n 3. Preservation of order
n 4. Smoothness of the disparity field

## Correspondence problem : methods

1. correlation
q deformations...
$q$ small window $\Rightarrow$ noise!
$q$ large window $\Rightarrow$ bad localisation
2. feature-based
q mainly edges and corners
q sparse depth image
3. regularisation methods

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## Stereo, the general setup

3D reconstruction

$$
\begin{aligned}
& P=C+\mu R K^{-1} p \\
& P=C^{\prime}+\mu^{\prime} R^{\prime} K^{\prime-1} p^{\prime}
\end{aligned}
$$

Yields 6 equations in 5 unknowns $\mathrm{X}, \mathrm{Y}, \mathrm{Z}$ and $\mu, \mu^{\prime}$

However, due to noise and errors, the rays may not intersect!
$\Rightarrow$ e.g. use the middle where the rays come closest

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3D city models - ground level
Mobile mapping example - for measuring


## Computer <br> 3D city models - ground level

Can also be turned into 3D for visualisation, but one needs to stay close to the camera viewpoints.

The example shown is of Quebec

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## Uncalibrated reconstruction

From 2 views...


If the camera translates...
An affine reconstruction can be made A projective reconstruction is always possible (if no pure rot.)

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## Uncalibrated reconstruction

From 3 general views taken with the same camera parameters...


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## Uncalibrated reconstruction



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## Uncalibrated reconstruction



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## Uncalibrated reconstruction - example



Univ. of Leuven

Computer Vision

## Shape-from-stills

# Input Images shots taken with Canon EOS D60 

(Resolution: 6,3 Megapixel )

## Computer Shape-from-stills

## www.arc3d.be

Webservice,
free for non-commercial use

Computer Vision

## 3D acquisition taxonomy



## Active triangulation

## POINT PROJECTED ON OBJECT

## THE STEREO CORRESPONDENCE PROBLEM CAN BE MADE TRIVIAL BY PROJECTING A POINT ONTO THE OBJECT SURFACE WITH A LASER. IT IS LIKE REPLACING ONE OF THE STEREO CAMERAS WITH A LASER.

## Active triangulation

INTERSECTION LASER RAY AND VIEWING RAY

INTERSECTION LASER RAY AND
, OBJECT SURFACE
,

LASER SPOT SEEN BY THE
CAMERA

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## Active triangulation



CAMERA'S CENTER OF PROJECFION

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## Active triangulation

LASER

Two lines do normally not intersect... Noise disrupts triangulation

LASER SPOT SEEN BY THE
CAMERA

NO INTERSECTION if SOME ERRORS IN,THE LINE EQS!


## Active triangulation

INTERSECTION LASER PLANE \& OBJECT SURFACE
LASER WITH CYLINDRICAL LENSE IN FRONT

## INTERSECTION

 LASER PLANE \& VIEWING RAY
## POINT ON THE

 LASER LINE SEEN BY THE CAMERACAMERA'S CENTER OF PROJEC〒 ION

## Active triangulation

INTERSECTION LASER PLANE \& OBJECT SURFACE
LASER WITH CYLINDRICAL LENSE IN FRONT

A plane and a line do normally intersect...
Noise has little influence on the triangulation

INTERSECTION LASER PLANE \& VIEWING RAY


CAMERA'S CENTER OF PROJECTION

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## Active triangulation



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## Active triangulation

Triangulation $\rightarrow$ 3D measurements


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## Active triangulation

Camera image


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## Active triangulation



Computer Vision

## Active triangulation

## Example 1 Cyberware laser scanners



Desktop model for small objects

Medium-sized objects
Body scanner
Head scanner


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## Active triangulation

## Example 2 Minolta



Portable desktop model

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## 3D acquisition taxonomy


patterns of a special shape are projected onto the scene
deformations of the patterns yield information on the shape

Focus is on combining a good resolution with a minimum number of pattern projections

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## Serial binary patterns

A sequence of patterns with increasingly fine subdivisions

Yields $2^{n}$ identifiable lines for only $n$ patterns


## Reducing the nmb of projections: colour

## Binary patterns

Yields $2^{\text {n }}$ identifiable lines for only $n$ patterns
Using colours, e.g. 3,
Yields $3^{\text {n }}$ identifiable lines for only $n$ patterns


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## One-shot implementation

3D from a single frame - KULeuven ‘96:


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One-shot implementation
KULeuven '81: checkerboard pattern with column code example :


# Computer Vision 

## 3D reconstruction for the example



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## An application in agriculture



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One-shot 3D acquisition

## Leuven ShapeCam



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Shape + texture often needed

Higher resolution
Texture is also extracted


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James Bond
Die another day

## Lara Croft

Thomb
Raider


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## Active triangulation

## Recent, commercial example



## KINECT

for 0 recurna
Kinect 3D camera, affordable and compact solution by Microsoft.

Projects a 2D point pattern in the NIR, to make it invisible to the human eye

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Kinect: $9 \times 9$ patches with locally unique code


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Kinect as one-shot, low-cost scanner

Excerpt from the dense NIR dot pattern:

http://research.microsoft.com/apps/video/default.aspx? 15

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## Face animation - input



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## Face animation - replay + effects



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## 4D: Facial motion capture

motion capture for League of Extraordinary Gentlemen


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## Facial motion capture

(2) COMPUTERCAFE


LC015 Eyetronics
1/ 291
$03 / 11 / 2003$

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## Phase shift

## color wheel



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## Phase shift

$$
\begin{aligned}
& I_{r}=A+R \cos (\phi-\theta) \\
& I_{g}=A+R \cos (\phi) \\
& I_{b}=A+R \cos (\phi+\theta)
\end{aligned}
$$

1. detect phase from 3 subsequently projected cosine patterns, shifted over 120 degrees
2. unwrap the phases / additional stereo
3. texture is obtained by summing the 3 images / color camera w. slower integration
$\begin{aligned} & \text { Computer } \\ & \text { Vision }\end{aligned}$
$A=\frac{I_{r}+I_{g}+I_{b}}{3}$
$\phi=\arctan \left(\tan \left(\frac{\theta}{2}\right) \frac{I_{r}-I_{b}}{2 I_{g}-I_{r}-I_{b}}\right)$


Vision

## 4D acquisition

Motion retargetting, from 3D phase shift scans


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## 3D acquisition taxonomy



## Time-of-flight

measurement of the time a modulated light signal needs to travel before returning to the sensor
this time is proportional to the distance
waves:

1. radar
2. sonar
3. optical radar
low freq. electromagnetic acoustic waves
optical waves
working principles :
4. pulsed
5. phase shifts

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Time-of-flight (optical radar /NIR)
Example 1: Cyrax

Example 2: Riegl


Vision

## Time-of-flight: example

## Cyrax ${ }^{\text {"w }}$

## 3D Laser Mapping

## System

Computer Vision

Cyrax

## Accurate, detailed, fast measuring



Integrated modeling

## Cyrax



Computer Vision

Pulsed laser (time-of-flight)
No reflectors needed

## 2mm-6mm accuracy

Distance $=C \times \Delta T \div 2$

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## Laser sweeps over surface

## 800 pts/sec



2mm min pt-to-pt spacing

## $40^{\circ} \times 40^{\circ}$

Field-of-view (max)

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## Up to 100m range (50m rec)

## Eye-safe Class 2

Computer Vision

Cyrax is also a visualization tool

Cyrax detects the intensity of each reflected laser pulse and colors it


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## Step 1:

## Target the structure

## Scanner Settings

Minimize Spacing

| $\left[\begin{array}{c}\text { Plane Offset Distance } \\ 200 \\ \mathrm{ft} \\ \hline\end{array}\right.$ |
| :--- |
| $\left[\begin{array}{r}\text { Spacing } \\ \text { Horizontal } \\ \text { Vertical } \\ \sqrt{0.2649} \\ \hline 0.08333\end{array} \mathrm{ft}\right.$ |


| Number of Points |
| :---: |
| Horizontal $\sqrt[500]{ }$ |
| Vertical |
| 500 |
| Scan |



Will scan 602000 points. Drag out target region with left mouse, or point with right mouse to set $t$

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Step 2:
Scan the structure


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Step 3: Color the points


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Step 4:
Model fitting in-the-field


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## Result




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## Project: As-built of

## Chevron hydrocarbon plant



- 400 ’x500' area
- 10 vessels; 5 pumps
- 6,000 objects
- 81 scans from 30 tripod locations
- Cyrax field time $=50 \mathrm{hrs}$

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## Added Value

 Benefits
## Measuring \& modeling



- Higher accuracy
- Fewer construction errors
- 6 week schedule savings


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Application Modeling movie sets


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## Lidar data with Riegl LMS-Z390i


courtesy of RWTH Aachen, L. Kobbelt et al.

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## Comparison Lidar - passive

## 3-D Reconstruction based on

## Multi-View Stereo

## LIDAR Measurements



# Computer 

 Vision
## 3D acquisition taxonomy



## Computer <br> Vision <br> Shape-from-texture

assumes a slanted and tilted surface to have a homogeneous texture
inhomogeneity is regarded as the result of projection
e.g. anisotropy in the statistics of edge orientations

## $\Downarrow$

orientations deprojecting to maximally isotropic texture

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Computer Vision

## 3D acquisition taxonomy



## Shape-from-contour

## makes assumptions about contour shape

E.g. the maximization of area over perimeter squared (compactness)

E.g. assumption of symmetry

Symmetric contours $\downarrow$ surface of revolution

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## Shape-from-contour



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## 3D acquisition taxonomy



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## Shape-from-silhouettes



Computer

## Shape from silhouettes - uncalibrated

 Vision
## tracking of turntable rotation

- volumetric modeling from silhouettes
- triangular textured surface mesh


Computer Vision

Shape from silhouettes - uncalibrated
tracking of turntable rotation

- volumetric modeling from silhouettes
- triangular textured surface mesh



## Computer

## Shape from silhouettes - uncalibrated

$\square$
VRML model

## Computer

Vision

## Computer <br> Outdoor shape-from-silhouettes

Vision


## Computer

Vision

## Outdoor shape-from-silhouettes


!
$\Rightarrow$


Computer Vision

Outdoor shape-from-silhouettes

E雷


Computer Vision

## 3D acquisition taxonomy



Computer Vision

# REAL-TIME FOCUS RANGE SENSOR 

SHQEE K. MAYAR<br>Wasahigo Watamaze<br>Minori Nocuchi<br>COLUMAIA UNIVERSITY

# Computer 

 Vision
## 3D acquisition taxonomy



## Shape-from-shading

Uses directional lighting, often with known direction
local intensity is brought into correspondence with orientation via reflectance maps
orientation of an isolated patch cannot be derived uniquely
extra assumptions on surface smoothness and known normals at the rim

Computer Vision

## 3D acquisition taxonomy



## Photometric stereo

constraint propagation eliminated by using light from different directions
simultaneously when the light sources are given different colours

## Computer Vision <br> Mini-dome for photometric stereo

Instead of working with multi-directional light applied simultaneously with the colour trick, one can also project from many directions in sequence...

Computer Vision

Mini-dome for photometric stereo

KATHOLIEKE UNIVERSITEIT


## Computer Vision <br> Mini-dome for photometric stereo

## Example for tablet with first world map known,

 an exhibit at the British Museum:http://homes.esat.kuleuven.be/~mproesma/mptmp/cuneiform

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## Mini-dome for photometric stereo



## Computer <br> 3D and recognition integrated

 Vision